

## STAT 610: Discussion 2

### 1 Summary

- Minimal sufficient statistics

- Definition: A statistic  $T(\mathbf{X})$  is minimal sufficient if for any other sufficient statistics  $T'(\mathbf{X})$ ,  $T(\mathbf{x})$  is a function of  $T'(\mathbf{x})$ .
- Theorem 6.2.13.  $T(\mathbf{X})$  is minimal sufficient for  $\boldsymbol{\theta}$  if  $T$  has the property that

$$T(\mathbf{x}) = T(\mathbf{y}) \Leftrightarrow \frac{f(\mathbf{x}|\boldsymbol{\theta})}{f(\mathbf{y}|\boldsymbol{\theta})} \text{ does not depend on } \boldsymbol{\theta}.$$

- Theorem 6.6.5. See lecture 3.

- In the case of exponential families

Let  $X_1, \dots, X_n$  be i.i.d. from an exponential family with pdf

$$f_{\boldsymbol{\theta}}(x) = h(x)c(\boldsymbol{\theta}) \exp\{\boldsymbol{\eta}(\boldsymbol{\theta})^T \mathbf{T}(x)\},$$

where  $\boldsymbol{\eta}(\boldsymbol{\theta})^T = (w_1(\boldsymbol{\theta}), \dots, w_k(\boldsymbol{\theta}))$ , and  $\mathbf{T}(x)^T = (t_1(x), \dots, t_k(x))$ . Then,

- $T(\mathbf{X}) = (\sum_{i=1}^n t_1(X_i), \dots, \sum_{i=1}^n t_k(X_i))$  is a sufficient statistic for  $\boldsymbol{\theta}$ .
- $T(\mathbf{X}) = (\sum_{i=1}^n t_1(X_i), \dots, \sum_{i=1}^n t_k(X_i))$  is also minimal sufficient if the parameter space contains an open set in  $\mathbb{R}^k$ .

- Ancillary statistics

- Definition:  $U(\mathbf{X})$  is ancillary of  $\theta$  if the distribution of  $U$  does not depend on  $\theta$ .

- Complete statistics

- Definition:  $T(\mathbf{X})$  is complete if  $T$  has the property that

$$\mathbb{E}_{\boldsymbol{\theta}} g(T) = 0 \text{ for any measurable } g \text{ and } \boldsymbol{\theta} \Rightarrow g(T) = 0 \text{ a.s.}$$

- Basu's Theorem: Complete sufficient statistics and ancillary statistics for parameter  $\theta$  are independent for all  $\theta$ .

## 2 Questions

1. Let  $(X_1, \dots, X_n)$  be a random sample from density  $\theta^{-1}e^{-(x-\theta)/\theta}I_{(\theta, \infty)}(x)$ , where  $\theta > 0$  is an unknown parameter.

- (a) Find a statistic that is minimal sufficient for  $\theta$ .
- (b) Show whether the minimal sufficient statistic in (a) is complete.

2. Let  $X$  be a discrete random variable with

$$\mathbb{P}_\theta(X = x) = \frac{\binom{\theta}{x} \binom{N-\theta}{n-x}}{\binom{N}{n}}, \quad x = 0, 1, 2, \dots, \min\{\theta, n\}, n - x \leq N - \theta$$

where  $n$  and  $N$  are known positive integers,  $N \geq n$ , and  $\theta = 0, 1, \dots, N$ . Show that  $X$  is complete.

3. (a) If  $\frac{X}{Y}$  and  $Y$  are independent random variables, show that

$$\mathbb{E} \left( \frac{X}{Y} \right)^k = \frac{\mathbb{E}(X^k)}{\mathbb{E}(Y^k)}.$$

- (b) Use Basu's theorem to show that if  $X_1, \dots, X_n$  are iid gamma( $\alpha, \beta$ ), where  $\alpha$  is known, then for  $T = \sum_{j=1}^n X_j$

$$\mathbb{E}(X_{(i)}|T) = \mathbb{E} \left( \frac{X_{(i)}}{T} T | T \right) = T \frac{\mathbb{E}(X_{(i)})}{\mathbb{E}T}$$

4. Let  $(X_1, \dots, X_n)$ ,  $n \geq 2$ , be a random sample from a distribution having density  $f_{\theta,j}$ , where  $\theta > 0$ ,  $j = 1, 2$ ,  $f_{\theta,1}$  is the density of  $\mathcal{N}(0, \theta^2)$ , and  $f_{\theta,2}(x) = (2\theta)^{-1}e^{-|x|/\theta}$ . Show that  $T = (T_1, T_2)$  is minimal sufficient for  $(\theta, j)$ , where  $T_1 = \sum_{i=1}^n X_i^2$  and  $T_2 = \sum_{i=1}^n |X_i|$ .